

```
import pandas as pd
import numpy as np
from sklearn.model_selection import KFold
import matplotlib.pyplot as plt
import sklearn
from sklearn.model_selection import train_test_split
from sklearn import datasets, linear_model
from sklearn.preprocessing import StandardScaler
from scipy.cluster.hierarchy import linkage, fcluster
from sklearn import metrics
import seaborn as sns
from sklearn.cluster import KMeans, DBSCAN
import plotly.figure_factory as ff
```

```
from google.colab import drive
drive.mount('/content/drive')
```

```
↳ Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).
```

```
# Import the CSV file
import pandas as pd
df = pd.read_csv("/content/drive/My Drive/FinalProject_ChicagoCrime/Merge_1.csv")
df.head()
```

```
↳
```

| | District | index | Case Number | Date | Block | IUCR | Primary Type | Description | Location Description | Arrest | Domestic | Beat | Ward | Community Area | FBI Code | X Coordinate | Y Coordinate |
|---|----------|-------|-------------|------------|--------------------------|------|---------------------|---------------------------|------------------------------|--------|----------|------|------|----------------|----------|--------------|--------------|
| 0 | 4 | 2 | JB528220 | 11/24/2018 | 095XX S STONY ISLAND AVE | 0261 | CRIM SEXUAL ASSAULT | AGGRAVATED: HANDGUN | ALLEY | 1 | 0 | 431 | 7.0 | 51.0 | 02 | 1188521.0 | 184211 |
| 1 | 22 | 3 | JB495848 | 08/27/2018 | 092XX S UNION AVE | 1120 | DECEPTIVE PRACTICE | FORGERY | RESIDENCE | 1 | 0 | 2223 | 21.0 | 73.0 | 10 | 1173264.0 | 184355 |
| 2 | 7 | 5 | JB505888 | 09/27/2018 | 068XX S PERRY AVE | 1020 | ARSON | BY FIRE | VEHICLE NON-COMMERCIAL | 1 | 0 | 722 | 6.0 | 69.0 | 09 | 1176575.0 | 185972 |
| 3 | 24 | 6 | JB472665 | 10/12/2018 | 061XX N BROADWAY | 0890 | THEFT | FROM BUILDING | COMMERCIAL / BUSINESS OFFICE | 1 | 0 | 2433 | 48.0 | 77.0 | 06 | 1167199.0 | 194057 |
| 4 | 8 | 9 | JB241744 | 04/27/2018 | 070XX S CALIFORNIA AVE | 1562 | SEX OFFENSE | AGG CRIMINAL SEXUAL ABUSE | APARTMENT | 0 | 1 | 831 | 17.0 | 66.0 | 17 | 1158921.0 | 185792 |

```
df['ArrestLabel']=df['Arrest']
df.head()
```

```
↳
```

| | District | index | Case Number | Date | Block | IUCR | Primary Type | Description | Location Description | Arrest | Domestic | Beat | Ward | Community Area | FBI Code | X Coordinate | Y Coordinate |
|---|----------|-------|-------------|------------|--------------------------|------|---------------------|---------------------------|------------------------------|--------|----------|------|------|----------------|----------|--------------|--------------|
| 0 | 4 | 2 | JB528220 | 11/24/2018 | 095XX S STONY ISLAND AVE | 0261 | CRIM SEXUAL ASSAULT | AGGRAVATED: HANDGUN | ALLEY | 1 | 0 | 431 | 7.0 | 51.0 | 02 | 1188521.0 | 184211.0 |
| 1 | 22 | 3 | JB495848 | 08/27/2018 | 092XX S UNION AVE | 1120 | DECEPTIVE PRACTICE | FORGERY | RESIDENCE | 1 | 0 | 2223 | 21.0 | 73.0 | 10 | 1173264.0 | 184355.0 |
| 2 | 7 | 5 | JB505888 | 09/27/2018 | 068XX S PERRY AVE | 1020 | ARSON | BY FIRE | VEHICLE NON-COMMERCIAL | 1 | 0 | 722 | 6.0 | 69.0 | 09 | 1176575.0 | 185972.0 |
| 3 | 24 | 6 | JB472665 | 10/12/2018 | 061XX N BROADWAY | 0890 | THEFT | FROM BUILDING | COMMERCIAL / BUSINESS OFFICE | 1 | 0 | 2433 | 48.0 | 77.0 | 06 | 1167199.0 | 194057.0 |
| 4 | 8 | 9 | JB241744 | 04/27/2018 | 070XX S CALIFORNIA AVE | 1562 | SEX OFFENSE | AGG CRIMINAL SEXUAL ABUSE | APARTMENT | 0 | 1 | 831 | 17.0 | 66.0 | 17 | 1158921.0 | 185792.0 |

```
#Dropping column
df=df.drop(columns=['Arrest'])
df.rename(columns={'ArrestLabel':'Arrest'},inplace=True)
df.head()
```

| | District | index | Case Number | Date | Block | IUCR | Primary Type | Description | Location Description | Domestic | Beat | Ward | Community Area | FBI Code | X Coordinate | Y Coordinate | Latitude |
|---|----------|-------|-------------|------------|--------------------------|------|---------------------|---------------------------|------------------------------|----------|------|------|----------------|----------|--------------|--------------|----------|
| 0 | 4 | 2 | JB528220 | 11/24/2018 | 095XX S STONY ISLAND AVE | 0261 | CRIM SEXUAL ASSAULT | AGGRAVATED: HANDGUN | ALLEY | 0 | 431 | 7.0 | 51.0 | 02 | 1188521.0 | 1842119.0 | 41.71 |
| 1 | 22 | 3 | JB495848 | 08/27/2018 | 092XX S UNION AVE | 1120 | DECEPTIVE PRACTICE | FORGERY | RESIDENCE | 0 | 2223 | 21.0 | 73.0 | 10 | 1173264.0 | 1843551.0 | 41.71 |
| 2 | 7 | 5 | JB505888 | 09/27/2018 | 068XX S PERRY AVE | 1020 | ARSON | BY FIRE | VEHICLE NON-COMMERCIAL | 0 | 722 | 6.0 | 69.0 | 09 | 1176575.0 | 1859723.0 | 41.71 |
| 3 | 24 | 6 | JB472665 | 10/12/2018 | 061XX N BROADWAY | 0890 | THEFT | FROM BUILDING | COMMERCIAL / BUSINESS OFFICE | 0 | 2433 | 48.0 | 77.0 | 06 | 1167199.0 | 1940579.0 | 41.91 |
| 4 | 8 | 9 | JB241744 | 04/27/2018 | 070XX S CALIFORNIA AVE | 1562 | SEX OFFENSE | AGG CRIMINAL SEXUAL ABUSE | APARTMENT | 1 | 831 | 17.0 | 66.0 | 17 | 1158921.0 | 1857929.0 | 41.71 |

```
df=df.drop(columns=['Beat','Ward','Latitude','Longitude','p_x','p_y','p_latitude','p_longitude','Case Number','Block','FBI Code','Date','p_dname','Primary Type','Se
df=df.drop(columns=["Description","Location Description","Domestic","Community Area","X Coordinate","Y Coordinate","Month","Day"])
```

```
df.info()
```

```

↳ <class 'pandas.core.frame.DataFrame'>
RangeIndex: 485344 entries, 0 to 485343
Data columns (total 7 columns):
District      485344 non-null int64
index         485344 non-null int64
IUCR          485344 non-null object
Distance      485344 non-null float64
Time          485344 non-null object
crimeGroups   485344 non-null int64
Arrest        485344 non-null int64
dtypes: float64(1), int64(4), object(2)
memory usage: 25.9+ MB

```

```

#making dummy variables out of the object variables
#df_dum= pd.get_dummies(df)
#df_dum.head()
#data['Ntime'] = (data['seconds'] - data['seconds'].min())/(data['seconds'].max()-data['seconds'].min())
df['IUCR Cat']=df['IUCR'].astype("category").cat.codes
X = df.drop(columns=["crimeGroups","Arrest","IUCR","Time"])
y = df[["crimeGroups"]]
print(X.shape, y.shape)

```

```

↳ (485344, 4) (485344, 1)

```

X.head

```

↳ <bound method NDFrame.head of
0      4      2  6822.245420      2
1     22      3 14717.784521     103
2      7      5  9500.832396      97
3     24      6  3987.355350      84
4      8      9  6432.894093     177
...     ...     ...     ...     ...
485339    8  224407  7287.873953     101
485340    7  224408  4288.521873     101
485341    7  224409  1541.306056     101
485342    5  224410  6056.942572     101
485343    6  224411  2708.275085     101

```

```

[485344 rows x 4 columns]>

```

```

y['crimeGroups'].unique()

```

```

↳ array([1, 3, 2, 4, 5])

```

```

#Dividing the data into Train and Test set
X_train, X_test, y_train, y_test = train_test_split(X,y, test_size = 0.2, random_state=0)

```

```

scaler = StandardScaler()
scaler.fit(X_train)
X_scaled = scaler.transform(X_train)

```

```

X_scaled[0:1]

```

```
↳ array([[ 0.83953522, -1.02231969, -1.03376793, -0.58829716]])
```

```
clustering = KMeans(n_clusters = 5, init = 'k-means++', n_init = 20, random_state=0).fit(X_scaled)
clusters = clustering.labels_
```

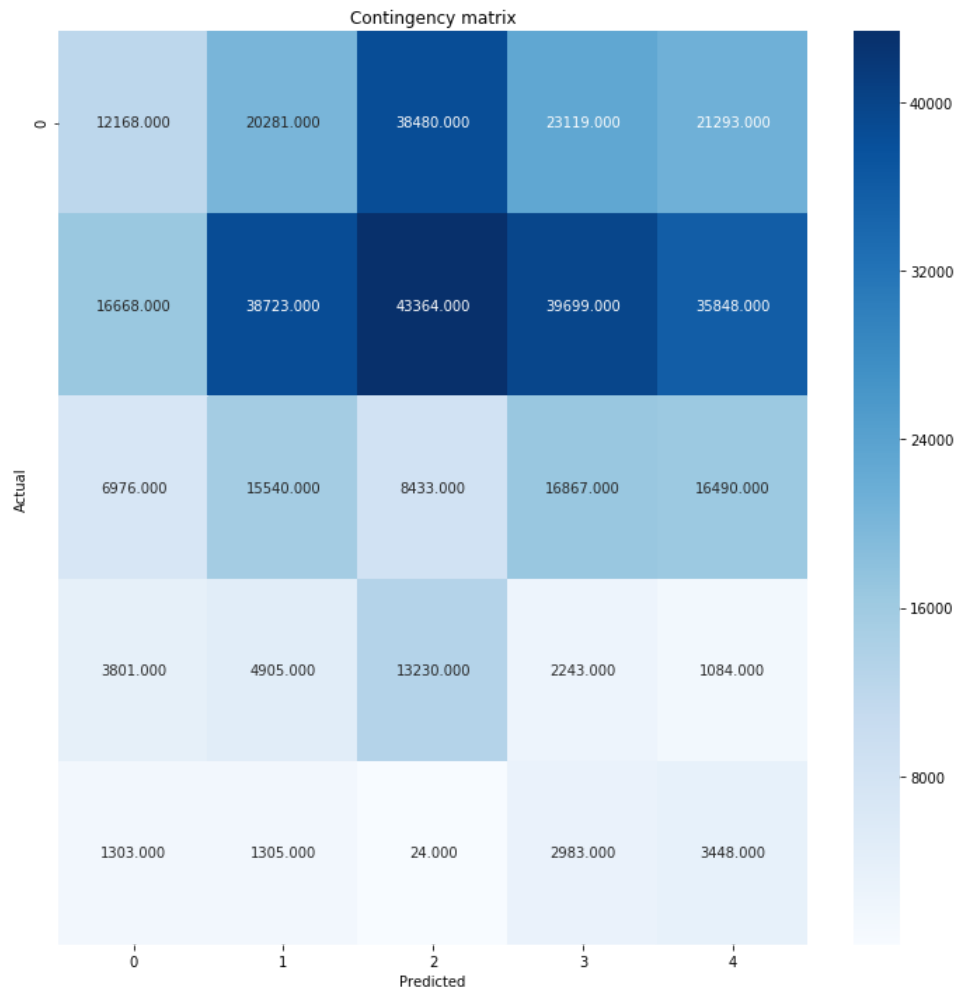
```
clusters[0:5]
```

```
↳ array([1, 3, 2, 4, 3], dtype=int32)
```

```
X_train['clusters'] = clusters - 1
```

```
cont_matrix = metrics.cluster.contingency_matrix(y_train, X_train['clusters'])
fig, ax = plt.subplots(figsize=(10,10))
ax=sns.heatmap(cont_matrix, yticklabels=5, annot = True, fmt = ".3f", square = False, cmap = plt.cm.Blues)
plt.ylabel('Actual')
plt.xlabel('Predicted')
plt.title('Contingency matrix')
plt.tight_layout()
```

```
↳
```



```
X_train['clusters'].shape
```

```
↳ (388275,)
```

```
Y=np.squeeze(y_train)
```

```
adjusted_rand_index = metrics.adjusted_rand_score(Y, X_train['clusters'])
silhouette_coefficient = metrics.silhouette_score(X_scaled, clusters, metric = "euclidean")
print([adjusted_rand_index, silhouette_coefficient])
```

```
↳ [0.01099136342906424, 0.22473666853860663]
```

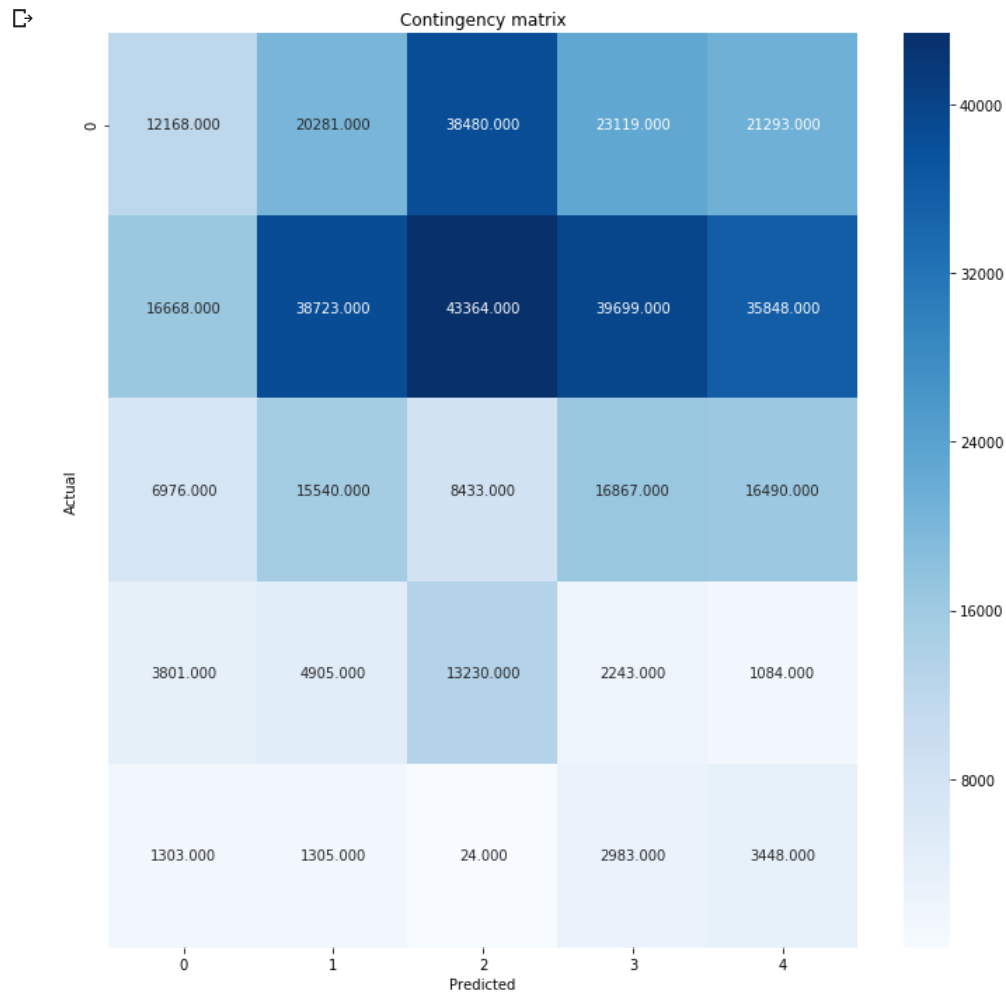
```
clustering = KMeans(n_clusters = 5, init = 'random', n_init = 10, random_state=0).fit(X_scaled)
clusters = clustering.labels_
```

```
clusters[0:5]
```

```
array([2, 1, 0, 3, 1], dtype=int32)
```

```
X_train['clusters'] = clusters - 1
```

```
cont_matrix = metrics.cluster.contingency_matrix(y_train,X_train['clusters'])  
fig, ax = plt.subplots(figsize=(10,10))  
ax=sns.heatmap(cont_matrix, yticklabels=5, annot = True, fmt = ".3f", square = False, cmap = plt.cm.Blues)  
plt.ylabel('Actual')  
plt.xlabel('Predicted')  
plt.title('Contingency matrix')  
plt.tight_layout()
```



```
X_train['clusters'].shape
```

(388275,)

```
Y=np.squeeze(y_train)
```

```
adjusted_rand_index = metrics.adjusted_rand_score(Y, X_train['clusters'])
silhouette_coefficient = metrics.silhouette_score(X_scaled, clusters, metric = "euclidean")
print([adjusted_rand_index, silhouette_coefficient])
```

[0.01127587904198441, 0.22532772840833132]

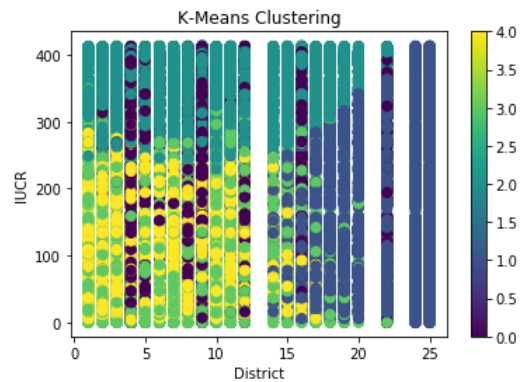
```
kmeans = pd.DataFrame(clusters)
```

```
kmeans[0]
```

```
0      1
1      3
2      2
3      4
4      3
..
388270  2
388271  0
388272  3
388273  2
388274  2
Name: 0, Length: 388275, dtype: int32
```

```
fig = plt.figure()
ax = fig.add_subplot(111)
scatter = ax.scatter(X_train['District'],X_train['IUCR Cat'],
                    c=kmeans[0],s=50)
ax.set_title('K-Means Clustering')
ax.set_xlabel('District')
ax.set_ylabel('IUCR')
plt.colorbar(scatter)
```

<matplotlib.colorbar.Colorbar at 0x7f0bad222ac8>



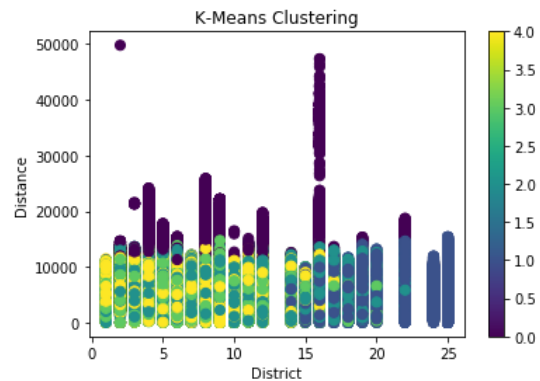
```
fig = plt.figure()
```

```

fig = plt.figure()
ax = fig.add_subplot(111)
scatter = ax.scatter(X_train['District'],X_train['Distance'],
                    c=kmeans[0],s=50)
ax.set_title('K-Means Clustering')
ax.set_xlabel('District')
ax.set_ylabel('Distance')
plt.colorbar(scatter)

```

☐ <matplotlib.colorbar.Colorbar at 0x7f0bab94e7b8>



```

clustering = DBSCAN(eps = 5, min_samples = 7, metric = "euclidean").fit(X_scaled)
clusters = clustering.labels_

cont_matrix = metrics.cluster.contingency_matrix(ytrain,clusters)
sns.heatmap(cont_matrix, annot = True, fmt = ".3f", square = True, cmap = plt.cm.Greens)
plt.ylabel('Actual')
plt.xlabel('Predicted')
plt.title('Contingency matrix')
plt.tight_layout()

adjusted_rand_index = metrics.adjusted_rand_score(ytrain, clusters)
silhouette_coefficient = metrics.silhouette_score(X_scaled, clusters, metric = "euclidean")
print([adjusted_rand_index, silhouette_coefficient])

```